Class 08 mini project

Jie

```
wisc.df <- read.csv("WisconsinCancer.csv", row.names=1)
head(wisc.df)</pre>
```

	diagnosis	radius_mean	texture_mean pe	erimeter_mean	area_mean	L
842302	M	17.99	10.38	122.80	1001.0)
842517	M	1 20.57	17.77	132.90	1326.0)
84300903	M	19.69	21.25	130.00	1203.0)
84348301	M	11.42	20.38	77.58	386.1	-
84358402	M	1 20.29	14.34	135.10	1297.0)
843786	M	12.45	15.70	82.57	477.1	-
	smoothnes	ss_mean compa	ctness_mean cond	cavity_mean co	oncave.poi	.nts_mean
842302	0	.11840	0.27760	0.3001		0.14710
842517	0	0.08474	0.07864	0.0869		0.07017
84300903	0	.10960	0.15990	0.1974		0.12790
84348301	0	14250	0.28390	0.2414		0.10520
84358402	0	0.10030	0.13280	0.1980		0.10430
843786	0	12780	0.17000	0.1578		0.08089
	symmetry_	mean fractal	_dimension_mean	radius_se tex	kture_se p	erimeter_se
842302	0.	2419	0.07871	1.0950	0.9053	8.589
842517	0.	1812	0.05667	0.5435	0.7339	3.398
84300903	0.	2069	0.05999	0.7456	0.7869	4.585
84348301	0.	2597	0.09744	0.4956	1.1560	3.445
84358402	0.	1809	0.05883	0.7572	0.7813	5.438
843786	0.	2087	0.07613	0.3345	0.8902	2.217
	area_se s	smoothness_se	${\tt compactness_se}$	concavity_se	concave.p	oints_se
842302	153.40	0.006399	0.04904	0.05373		0.01587
842517	74.08	0.005225	0.01308	0.01860		0.01340
84300903	94.03	0.006150	0.04006	0.03832		0.02058
84348301	27.23	0.009110	0.07458	0.05661		0.01867
84358402	94.44	0.011490	0.02461	0.05688		0.01885

843786	27.19	0.00751	0	0.033	45 0.03	3672	0.01137
	symmetry_se	fractal_	dimens	ion_se r	adius_worst	texture_worst	
842302	0.03003		0.0	006193	25.38	17.33	
842517	0.01389		0.0	003532	24.99	23.41	
84300903	0.02250		0.0	004571	23.57	25.53	
84348301	0.05963		0.0	009208	14.91	26.50	
84358402	0.01756		0.0	005115	22.54	16.67	
843786	0.02165		0.0	005082	15.47	23.75	
	perimeter_wor	rst area	_worst	smoothn	ess_worst co	ompactness_wor	st
842302	184	.60	2019.0		0.1622	0.66	56
842517	158	.80	1956.0		0.1238	0.18	66
84300903	152	.50	1709.0		0.1444	0.42	45
84348301	98	. 87	567.7		0.2098	0.86	63
84358402	152	. 20	1575.0		0.1374	0.20	50
843786	103	.40	741.6		0.1791	0.52	49
	concavity_wor	rst conc	ave.po:	ints_wor	st symmetry	_worst	
842302	0.73	119		0.26	54 (0.4601	
842517	0.24	416		0.18	60 (0.2750	
84300903	0.49	504		0.24	30 (0.3613	
84348301	0.68	369		0.25	75 (0.6638	
84358402	0.40	000		0.16	25 (0.2364	
843786	0.53	355		0.17	41 (0.3985	
	fractal_dimen	nsion_wo	rst				
842302		0.11	890				
842517		0.08	902				
84300903		0.08	758				
84348301		0.17	300				
84358402		0.07	678				
843786		0.12	440				

Q1. How many observations are in this dataset? Q2. How many of the observations have a malignant diagnosis?

```
wisc.data <- wisc.df[,-1]
diagnosis <- as.factor(wisc.df$diagnosis)
length(diagnosis)</pre>
```

[1] 569

```
table(diagnosis)
```

```
diagnosis
 В
     Μ
357 212
Q3. How many variables/features in the data are suffixed with _mean?
  grep("_mean",colnames(wisc.data))
 [1] 1 2 3 4 5 6 7 8 9 10
  colnames(wisc.data)
 [1] "radius_mean"
                                "texture_mean"
 [3] "perimeter_mean"
                                "area_mean"
 [5] "smoothness_mean"
                                "compactness_mean"
 [7] "concavity mean"
                                "concave.points_mean"
 [9] "symmetry_mean"
                                "fractal_dimension_mean"
[11] "radius se"
                                "texture se"
[13] "perimeter_se"
                                "area se"
[15] "smoothness_se"
                                "compactness_se"
[17] "concavity_se"
                                "concave.points_se"
[19] "symmetry_se"
                                "fractal_dimension_se"
[21] "radius_worst"
                                "texture_worst"
[23] "perimeter_worst"
                                "area_worst"
[25] "smoothness_worst"
                                "compactness_worst"
[27] "concavity_worst"
                                "concave.points_worst"
                                "fractal_dimension_worst"
[29] "symmetry_worst"
  colMeans(wisc.data)
            radius mean
                                    texture mean
                                                           perimeter_mean
           1.412729e+01
                                    1.928965e+01
                                                             9.196903e+01
              area mean
                                 smoothness mean
                                                         compactness_mean
           6.548891e+02
                                    9.636028e-02
                                                             1.043410e-01
         concavity_mean
                            concave.points_mean
                                                            symmetry_mean
```

4.891915e-02

4.051721e-01

radius_se

1.811619e-01

1.216853e+00

texture_se

8.879932e-02

6.279761e-02

fractal_dimension_mean

${\tt smoothness_se}$	area_se	perimeter_se
7.040979e-03	4.033708e+01	2.866059e+00
concave.points_se	concavity_se	compactness_se
1.179614e-02	3.189372e-02	2.547814e-02
radius_worst	${\tt fractal_dimension_se}$	symmetry_se
1.626919e+01	3.794904e-03	2.054230e-02
area_worst	perimeter_worst	texture_worst
8.805831e+02	1.072612e+02	2.567722e+01
concavity_worst	compactness_worst	smoothness_worst
2.721885e-01	2.542650e-01	1.323686e-01
${\tt fractal_dimension_worst}$	symmetry_worst	concave.points_worst
8.394582e-02	2.900756e-01	1.146062e-01

apply(wisc.data,2,sd)

radius_mean	texture_mean	perimeter_mean
3.524049e+00	4.301036e+00	2.429898e+01
area_mean	${\tt smoothness_mean}$	compactness_mean
3.519141e+02	1.406413e-02	5.281276e-02
concavity_mean	concave.points_mean	symmetry_mean
7.971981e-02	3.880284e-02	2.741428e-02
fractal_dimension_mean	radius_se	texture_se
7.060363e-03	2.773127e-01	5.516484e-01
perimeter_se	area_se	smoothness_se
2.021855e+00	4.549101e+01	3.002518e-03
compactness_se	concavity_se	concave.points_se
1.790818e-02	3.018606e-02	6.170285e-03
symmetry_se	fractal_dimension_se	radius_worst
8.266372e-03	2.646071e-03	4.833242e+00
texture_worst	perimeter_worst	area_worst
6.146258e+00	3.360254e+01	5.693570e+02
smoothness_worst	compactness_worst	concavity_worst
2.283243e-02	1.573365e-01	2.086243e-01
<pre>concave.points_worst</pre>	symmetry_worst	<pre>fractal_dimension_worst</pre>
6.573234e-02	6.186747e-02	1.806127e-02

Practical PCA issue: Scalinng

```
wisc.pca <- prcomp(wisc.data,scale=TRUE)
summary(wisc.pca)</pre>
```

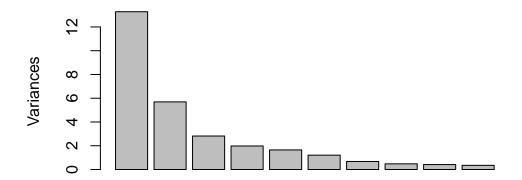
Importance of components:

```
PC1
                                 PC2
                                          PC3
                                                  PC4
                                                          PC5
                                                                  PC6
                                                                          PC7
Standard deviation
                       3.6444 2.3857 1.67867 1.40735 1.28403 1.09880 0.82172
Proportion of Variance 0.4427 0.1897 0.09393 0.06602 0.05496 0.04025 0.02251
                       0.4427\ 0.6324\ 0.72636\ 0.79239\ 0.84734\ 0.88759\ 0.91010
Cumulative Proportion
                           PC8
                                   PC9
                                          PC10
                                                 PC11
                                                         PC12
                                                                 PC13
                                                                         PC14
Standard deviation
                       0.69037 0.6457 0.59219 0.5421 0.51104 0.49128 0.39624
Proportion of Variance 0.01589 0.0139 0.01169 0.0098 0.00871 0.00805 0.00523
Cumulative Proportion
                       0.92598 0.9399 0.95157 0.9614 0.97007 0.97812 0.98335
                                           PC17
                                                   PC18
                                                           PC19
                                                                   PC20
                          PC15
                                   PC16
                                                                           PC21
Standard deviation
                       0.30681 0.28260 0.24372 0.22939 0.22244 0.17652 0.1731
Proportion of Variance 0.00314 0.00266 0.00198 0.00175 0.00165 0.00104 0.0010
Cumulative Proportion
                       0.98649 0.98915 0.99113 0.99288 0.99453 0.99557 0.9966
                          PC22
                                   PC23
                                          PC24
                                                  PC25
                                                          PC26
                                                                  PC27
Standard deviation
                       0.16565 0.15602 0.1344 0.12442 0.09043 0.08307 0.03987
Proportion of Variance 0.00091 0.00081 0.0006 0.00052 0.00027 0.00023 0.00005
Cumulative Proportion 0.99749 0.99830 0.9989 0.99942 0.99969 0.99992 0.99997
                          PC29
                                   PC30
Standard deviation
                       0.02736 0.01153
Proportion of Variance 0.00002 0.00000
Cumulative Proportion 1.00000 1.00000
```

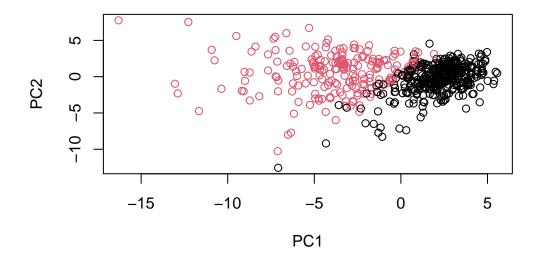
- Q4. From your results, what proportion of the original variance is captured by the first principal components (PC1)? 44%
- Q5. How many principal components (PCs) are required to describe at least 70% of the original variance in the data? 28
- Q6. How many principal components (PCs) are required to describe at least 90% of the original variance in the data? 24

```
plot(wisc.pca)
```

wisc.pca



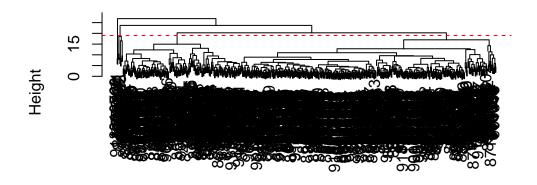
plot(wisc.pca\$x[,1],wisc.pca\$x[,2],xlab="PC1",ylab="PC2",col=diagnosis)



```
data.scaled <- scale(wisc.data)
wisc.hclust <- hclust(dist(data.scaled))
plot(wisc.hclust)

#grps <- cutree(wisc.hclust,k=1)
#abline(h=grps,col="red",lty=2)
abline(h=19,col="red",lty=2)</pre>
```

Cluster Dendrogram



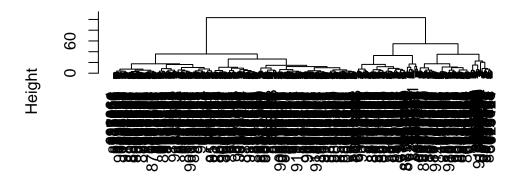
dist(data.scaled) hclust (*, "complete")

combine method

```
Our PCA results
```

```
wisc.pca.hclust <- hclust(dist(wisc.pca$x[,1:3]), method = "ward.D2")
plot(wisc.pca.hclust)</pre>
```

Cluster Dendrogram



dist(wisc.pca\$x[, 1:3]) hclust (*, "ward.D2")

```
grps <- cutree(wisc.pca.hclust,k=2)
table(grps)

grps
1    2
203 366

table(grps, diagnosis)

diagnosis
grps    B    M
1    24 179
2    333    33

#library(rgl)
#plot3d(wisc.pca$x[,1:3], xlab="PC 1", ylab="PC 2", zlab="PC 3", cex=1.5, size=1, type="s"</pre>
```